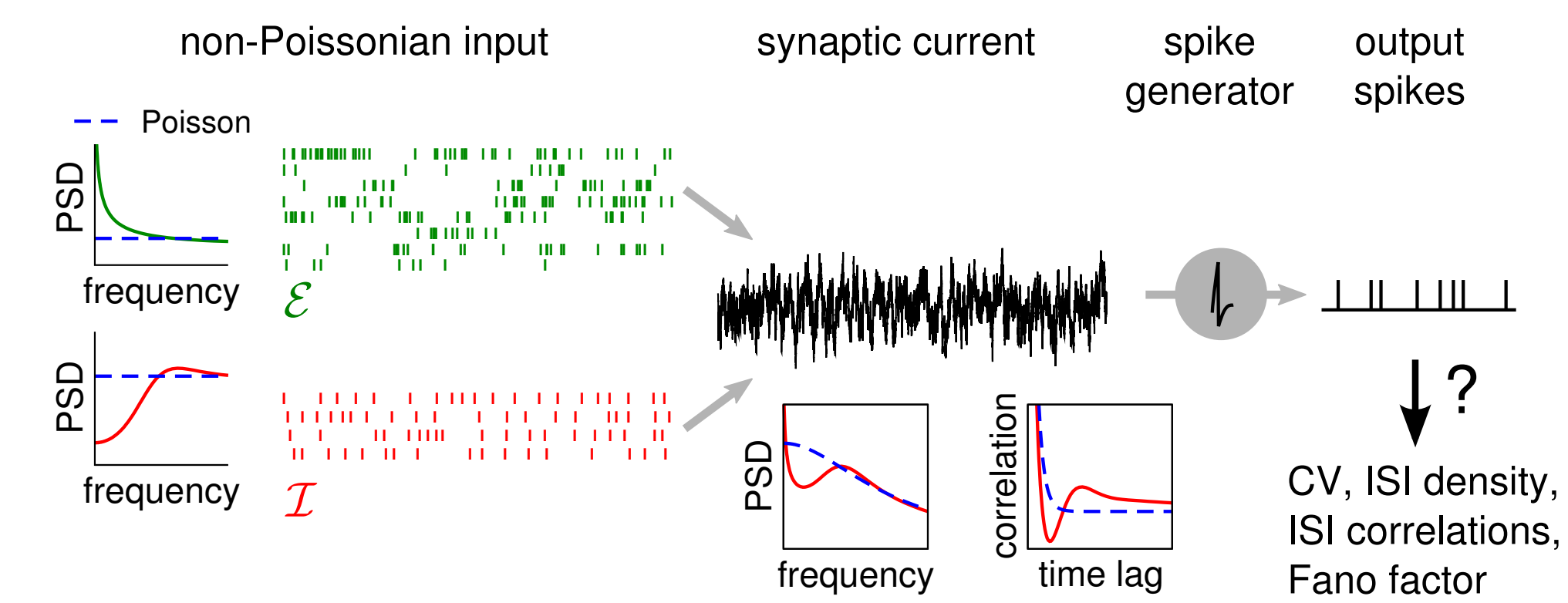


Statistical structure of neural spiking under non-Poissonian stimulation

Motivation

- Can we understand the interspike interval (ISI) statistics of spontaneous neural activity?
- What is the relation between input and output statistics of a neuron? --> Important for understanding population activity.
- Most theoretical studies assume that neurons are driven by Poisson spike trains ("white noise" processes, i.e. uncorrelated in time)
- However, realistic synaptic inputs have temporal structure ("colored noise"), e.g. due to refractoriness, bursting, structured "signals", short-term synaptic plasticity, oscillatory activity, adaptation.



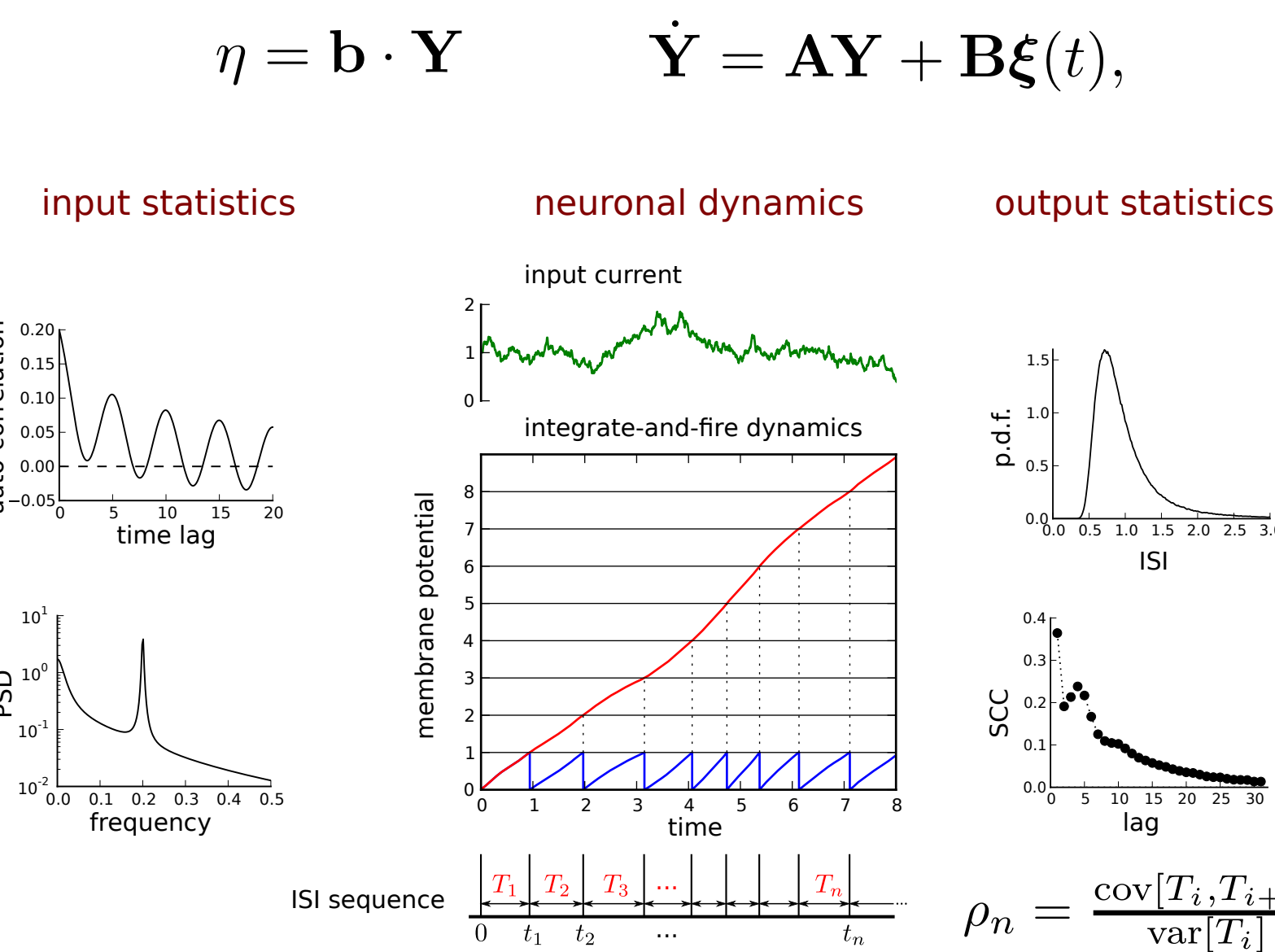
Wanted:
Spiking statistics for *arbitrary input correlation function*?

General theory

Perfect integrate-and-fire model

$$C_m \dot{V} = I_0 + I_{\text{syn}}(t), \quad \text{if } V = V_{\text{th}}: \quad V \rightarrow 0.$$

- diffusion approximation: $\dot{V} = \mu + \sigma \eta(t)$
- noise correlation function $C(t) \equiv \langle \eta(t') \eta(t' + t) \rangle$
- first-passage problem* for non-Markovian process $V(t)$
- Trick:** d-dimensional Ornstein-Uhlenbeck process



Fokker-Planck equation for $p(v, \mathbf{y}, t)$

$$\frac{\partial p}{\partial t} = -(\mu + \sigma \sum_{i=1}^d b_i y_i) \frac{\partial p}{\partial v} - \sum_{i,j=1}^d A_{ij} \frac{\partial}{\partial y_i} (y_j p) + \sum_{i,j=1}^d D_{ij} \frac{\partial^2 p}{\partial y_i \partial y_j}$$

$$\text{initial condition: } p(v, \mathbf{y}, 0) = \delta(v) \frac{1 + \epsilon \mathbf{b} \cdot \mathbf{y}}{\sqrt{(2\pi)^d |\Sigma|}} \exp(-\frac{1}{2} \mathbf{y}^T \Sigma \mathbf{y})$$

n-th-order interval density

$$P_n(t) = \int d^d \mathbf{y} \underbrace{J_v(nV_{\text{th}}, \mathbf{y}, t)}_{(\mu + \sigma \mathbf{b} \cdot \mathbf{y}) p(nV_{\text{th}}, \mathbf{y}, t)} \underbrace{q}_{q = \langle e^{i\ell V + i\mathbf{k}^T \mathbf{Y}} \rangle}$$

$$= \int \frac{d\ell}{2\pi} e^{-inV_{\text{th}}\ell} (\mu - i\sigma \mathbf{b}^T \nabla_{\mathbf{k}}) q(\ell, \mathbf{k}, t)|_{\mathbf{k}=0}$$

moment generating function

$$\bar{P}_n(s) = \int_0^\infty dt e^{-st} P_n(t) = \mu(1 - i\epsilon \mathbf{b}^T \nabla_{\mathbf{k}}) \varphi(nV_{\text{th}}, \mathbf{k}, s)|_{\mathbf{k}=0}$$

$$\varphi(v, \mathbf{k}, s) = \int_0^\infty dt e^{-st} \int d^d \mathbf{y} e^{i\mathbf{k} \cdot \mathbf{y}} p(v, \mathbf{y}, t)$$

$$\text{weak-noise perturbation theory: } \varphi = \varphi^{(0)} + \epsilon \varphi^{(1)} + \epsilon^2 \varphi^{(2)} + \dots$$

Analytical results

n^{th} -order ISI density

$$P_n(t) = \frac{1}{2\langle T \rangle \sqrt{4\pi\epsilon^2 h^3(t)}} \exp \left[-\frac{(t - n\langle T \rangle)^2}{4\epsilon^2 \langle T \rangle^2 h(t)} \right] \times \left\{ \frac{\left[\left(n - \frac{t}{\langle T \rangle} \right) g(t) + 2h(t) \right]^2}{2h(t)} - \epsilon^2 [g^2(t) - 2h(t)C(t)] \right\}$$

$$\langle T \rangle = V_{\text{th}}/\mu \quad g(t) = \frac{1}{\langle T \rangle} \int_0^t dt' C(t') \quad h(t) = \frac{1}{\langle T \rangle} \int_0^t dt' g(t')$$

ISI cumulants

$$\kappa_{2,n} = 2\langle T \rangle^2 \epsilon^2 \{ h_n + \epsilon^2 [g_n^2 + C(n\langle T \rangle)h_n] \} \quad g_n = g(n\langle T \rangle)$$

$$\kappa_{3,n} = 12\langle T \rangle^3 \epsilon^4 g_n h_n, \quad h_n = h(n\langle T \rangle)$$

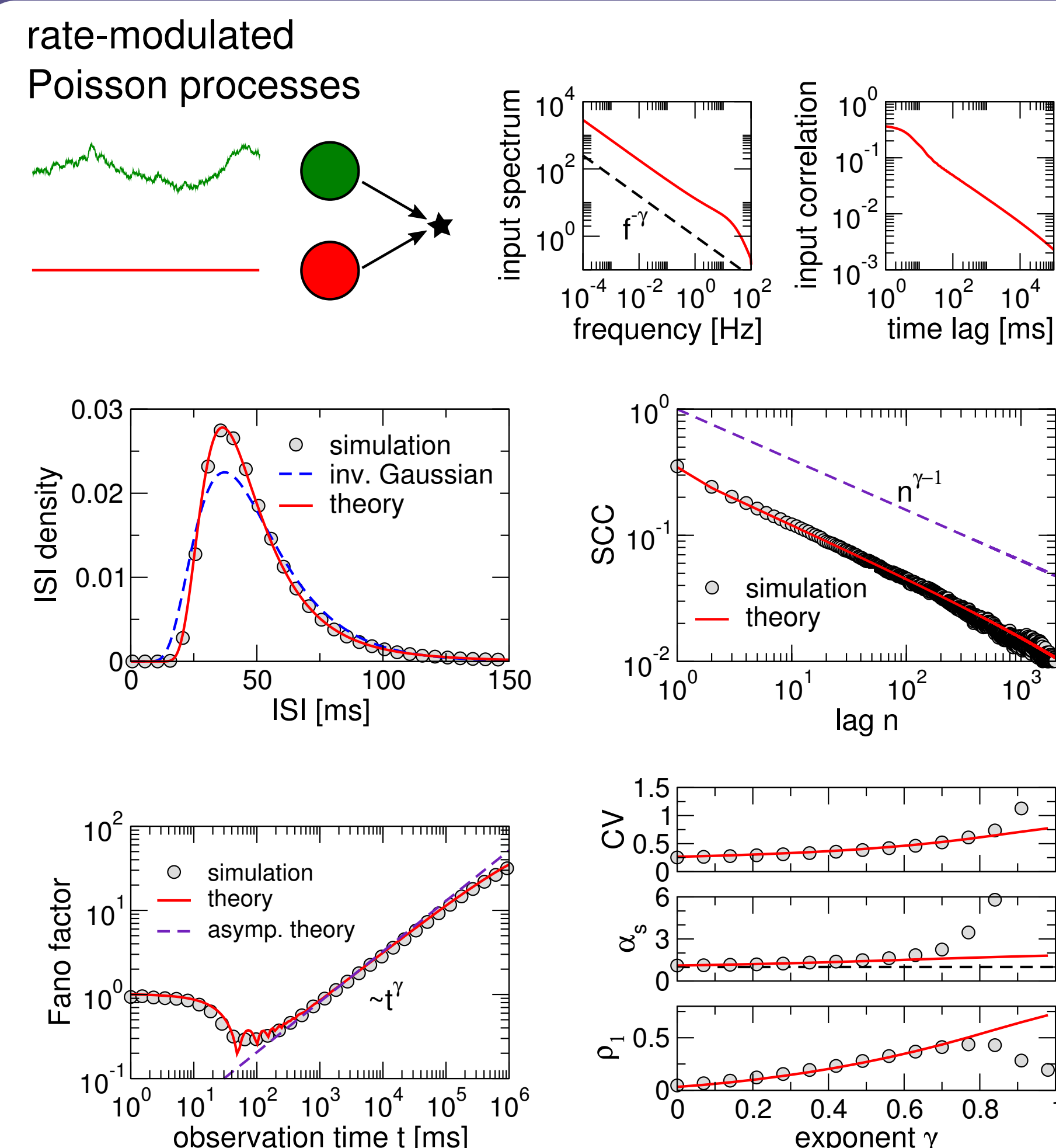
CV, skewness and ISI serial correlations

$$C_V^2 \approx 2\epsilon^2 h(\langle T \rangle) = \epsilon^2 \int df \tilde{C}(f) \text{sinc}^2(f\langle T \rangle) \quad \text{sinc}(x) = \frac{\sin(\pi x)}{\pi x}$$

$$\alpha_s \approx \frac{g(\langle T \rangle)}{h(\langle T \rangle)} = \frac{2 \int df \tilde{C}(f) \text{sinc}(2f\langle T \rangle)}{\int df \tilde{C}(f) \text{sinc}^2(f\langle T \rangle)}$$

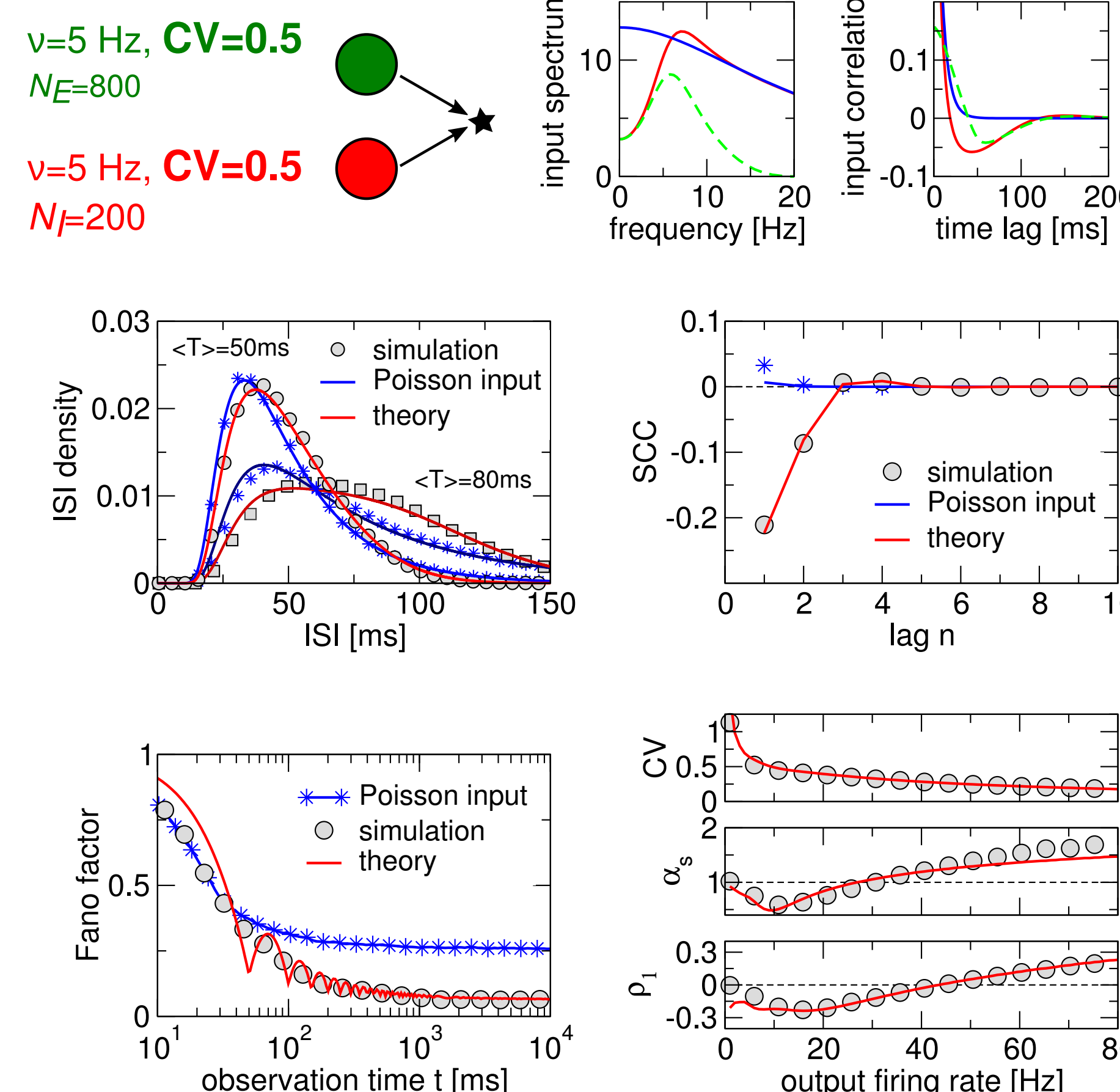
$$\rho_n = \frac{\kappa_{2,n+1} - 2\kappa_{2,n} + \kappa_{2,n-1}}{2\kappa_{2,1}} \approx \frac{\int df \tilde{C}(f) \text{sinc}^2(f\langle T \rangle) e^{-2\pi i f \langle T \rangle n}}{\int df \tilde{C}(f) \text{sinc}^2(f\langle T \rangle)}$$

Powerlaw input

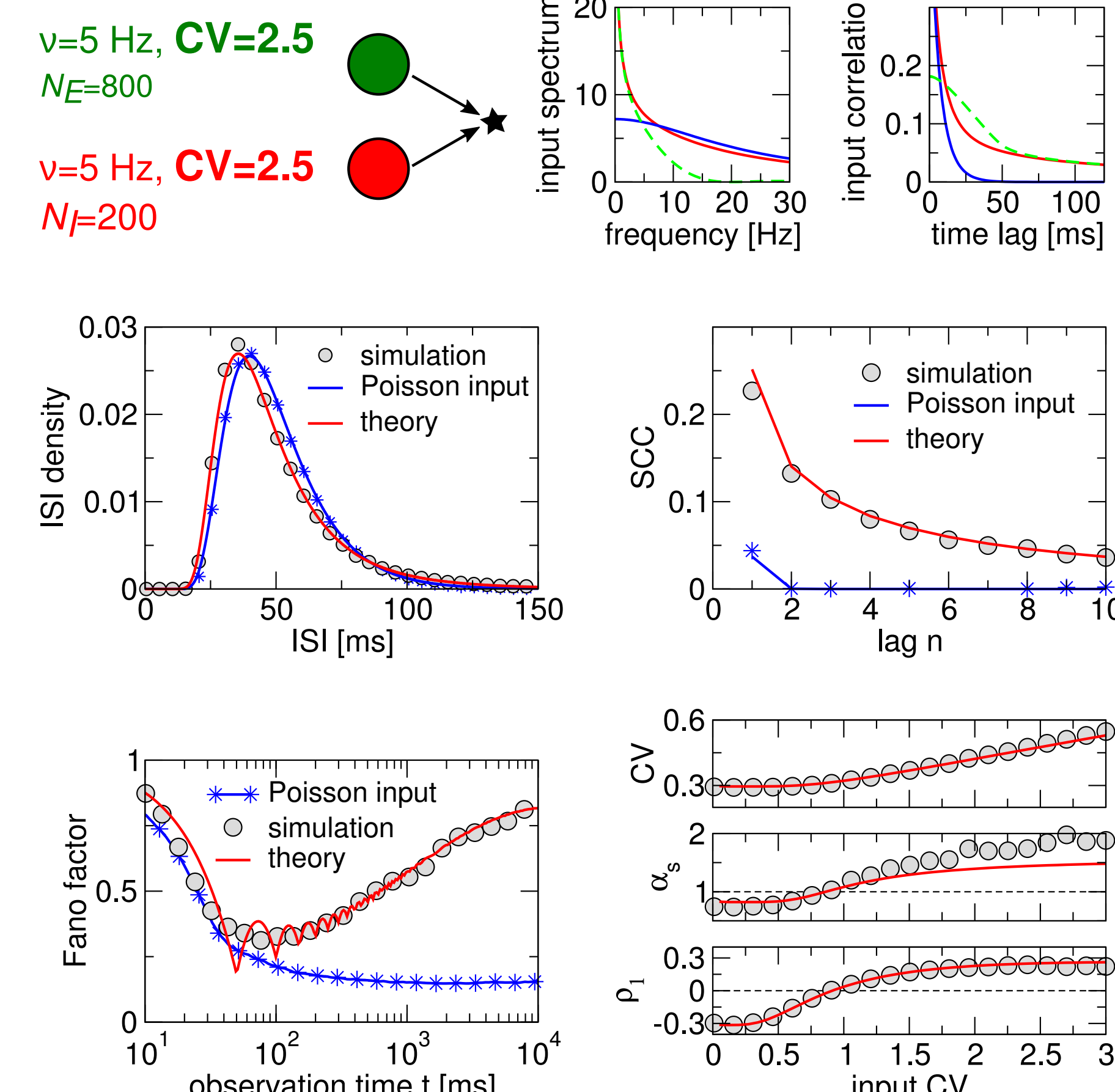


Non-Poissonian input

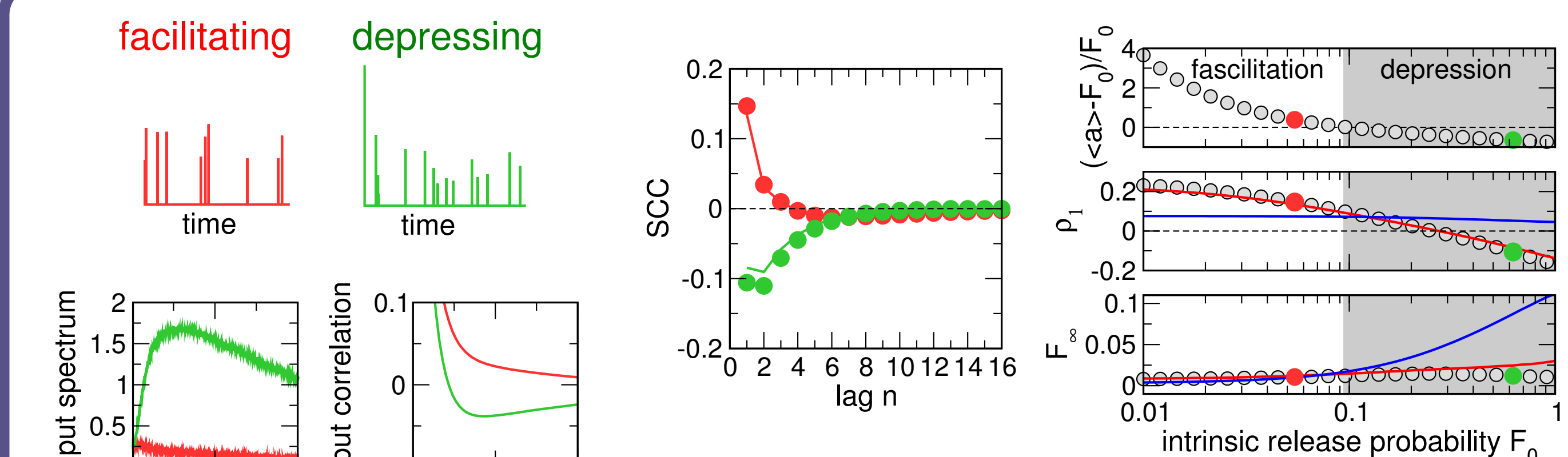
regular input spikes (refractoriness)



irregular input spikes (burstiness)



Short-term synaptic plasticity



Conclusions

- For Gaussian input with arbitrary correlation function, we derived analytical formulas for ISI density, auto-correlogram, CV, skewness, ISI serial correlations and Fano factor.
- works well for mean-driven regime with CV~0.5 or smaller
- regular input or synaptic depression => negative ISI correlations, small skew
- irregular input or facilitation => positive ISI correlations, large skew
- scaling behaviors for powerlaw input